

**Algorithm 1** Preprocessing of benchmark dataset

1: **Input:** A set of audio files  $\mathcal{F}_\mu = \{f_1, f_2, \dots, f_n\}$  with labels  $\mathcal{L} = \{\text{HC}, \text{PD}\}$ ; segment duration  $T_s$  (in seconds); sampling rate  $f_s$  (Hz); noise duration  $T_{\text{noise}}$ ; spectral threshold coefficient  $\alpha$

2: **Output:** A set of segmented audio signals  $\mathcal{Y}_{\text{segments}}$  for each processed file

3: **Step 1: Data Preprocessing (Noise Reduction)**

4: **for** each audio file  $f \in \mathcal{F}_\mu$  **do**

5:   Load the audio file:  $(y(t), f_s) \leftarrow \text{load}(f)$

6:   Sample the speech signal at  $f_s$  to obtain discrete samples  $y[n]$

7:   **Spectral Gating:**

8:   Compute the Short-Time Fourier Transform (STFT):

$$Y[k, m] = \text{STFT}(y[n]) = \sum_n y[n] \cdot w[n - m] e^{-j2\pi kn/N}$$

9:   Estimate the noise spectrum from the initial segment  $y_{\text{noise}}[n]$  (duration  $T_{\text{noise}}$ ):

$$N[k] = \frac{1}{M} \sum_{m=1}^M |Y[k, m]|$$

10:   Apply spectral threshold  $T[k] = \alpha N[k]$ , where  $\alpha$  is the threshold ratio

11:   **for** each frequency bin  $k$  and time frame  $m$  **do**

12:     Perform spectral gating:

$$Y_{\text{denoised}}[k, m] = \begin{cases} Y[k, m], & \text{if } |Y[k, m]| \geq T[k] \\ 0, & \text{otherwise} \end{cases}$$

13:   **end for**

14:   Compute the Inverse STFT to obtain the denoised signal:

$$y_{\text{denoised}}[n] = \text{ISTFT}(Y_{\text{denoised}}[k, m])$$

15: **end for**

16: **Step 2: Audio Segmentation**

17: **for** each denoised signal  $y_{\text{denoised}}^i[n] \in \mathcal{Y}_{\text{denoised}}$  **do**

18:   Compute segment length  $S = T_s \times f_s$

19:   Divide the signal into  $L = \left\lfloor \frac{N}{S} \right\rfloor$  segments

20:   **for** each segment  $j = 1$  to  $L$  **do**

21:     Compute start and end indices:

$$\text{start}_j = (j - 1) \times S, \quad \text{end}_j = j \times S$$

22:     Extract the segment:

$$y_j[n] = y_{\text{denoised}}^i[n], \quad n \in [\text{start}_j, \text{end}_j)$$

23:     Append  $y_j[n]$  to  $\mathcal{Y}_{\text{segments}}$

24:   **end for**

25: **end for**

26: **return** Segmented audio signals  $\mathcal{Y}_{\text{segments}}$

**Algorithm 2** Tabular Feature Extraction

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- 1: **Input:** Denoised audio signals  $\mathcal{Y}_{\text{segments}} = \{y_{\text{denoised}}^1, y_{\text{denoised}}^2, \dots, y_{\text{denoised}}^n\}$ ; segment duration  $T_s$ ; sampling rate  $f_s$
  - 2: **Output:** Feature matrix  $X$  with extracted features for each segment
  - 3: **Step 2: Compute Features for Each Segment**
  - 4: **for** each segment  $y_j[n] \in \mathcal{Y}_{\text{segments}}$  **do**
  - 5:     **Statistical Features:**
  - 6:     Compute energy:

$$E_j = \sqrt{\frac{1}{S} \sum_{n=1}^S y_j[n]^2}$$

- 7:     Compute entropy of energy:

$$H_j = - \sum_{n=1}^S p_n \log(p_n), \quad p_n = \frac{y_j[n]^2}{\sum_{n=1}^S y_j[n]^2}$$

- 8:     Compute kurtosis and skewness:

$$\text{Kurtosis}_j = \frac{\frac{1}{S} \sum_{n=1}^S (y_j[n] - \mu_j)^4}{\sigma_j^4} - 3$$

$$\text{Skewness}_j = \frac{\frac{1}{S} \sum_{n=1}^S (y_j[n] - \mu_j)^3}{\sigma_j^3}$$

- 9:     **Praat Features:**
- 10:     Compute jitter, shimmer, fundamental frequency ( $F_0$ ), and harmonic-to-noise ratio (HNR):

$$J_{\text{local},j}, S_{\text{local},j}, F_{0,j}, \text{HNR}_j$$

- 11:     **Spectral Features:**
- 12:     Compute zero-crossing rate:

$$\text{ZCR}_j = \frac{1}{S-1} \sum_{n=1}^{S-1} |\text{sgn}(y_j[n]) - \text{sgn}(y_j[n+1])|$$

- 13:     Compute spectral centroid:

$$\text{SC}_j = \frac{\sum_k k \cdot |Y_j[k]|}{\sum_k |Y_j[k]|}$$

- 14:     Compute Mel-Frequency Cepstral Coefficients (MFCCs):

$$\text{MFCCs}_j = \text{DCT}(\log(\text{MelSpectrum}(y_j[n])))$$

- 15:     Append features  $[E_j, H_j, \text{Kurtosis}_j, \text{Skewness}_j, J_{\text{local},j}, S_{\text{local},j}, F_{0,j}, \text{HNR}_j, \text{ZCR}_j, \text{SC}_j, \text{MFCCs}_j]$  to feature matrix  $X$
  - 16:     **end for**
  - 17:     **return** Feature matrix  $X$
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**Algorithm 3** Speech Signal to Spectrogram

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1: Input: Audio files  $F_\mu = \{f_1, f_2, \dots, f_n\}$ , segment duration  $T_s = 5$  seconds
2: Output: Spectrogram images (STFT, Mel, MFCC, CQT, Chroma)
3: Step 1: Audio Segmentation
4: for each file  $f \in F_\mu$  do
5:   Load audio file  $y$ ,  $sr = \text{load}(f)$ 
6:   Compute segment length  $S = T_s \times sr$ 
7:   for each chunk  $i$  from 1 to  $\left\lfloor \frac{N}{S} \right\rfloor$  do
8:     Extract chunk  $y_i = y[\text{start}_i : \text{end}_i]$ 
9:   end for
10: end for
11: Step 2: Noise Reduction Using Spectral Gating
12: for each chunk  $y_i$  do
13:   Extract noise profile  $y_{\text{noise}} = y[0 : N_p]$ 
14:   Compute STFT of noise profile  $M_{\text{noise}}$ 
15:   Apply spectral threshold to filter  $M_{\text{denoised}}$ 
16: end for
17: Step 3: Spectrogram Generation
18: for each denoised chunk  $y_{\text{denoised}}$  do
19:   Compute STFT, Mel-spectrogram, MFCC, CQT, Chroma
20: end for
21: Step 4: Save Spectrograms
22: for each spectrogram type  $S$  do
23:   Save corresponding image to output directory
24: end for

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